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# Darshan: Enabling Application I/O Understanding in an Evolving HPC Landscape

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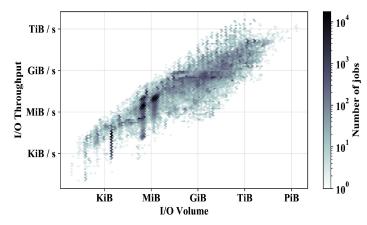


Figure courtesy of Mihailo Isakov (TAMU)



## **Understanding and improving HPC I/O**

- The ability to characterize and understand application I/O workloads is critical to ensuring efficient use of an evolving and increasingly complex HPC I/O stack
  - Deep layers of coordinating I/O libraries and entirely new-to-HPC storage paradigms (e.g., object storage)
  - Emerging storage hardware (e.g., PMEM) and storage architectures (e.g., burst buffers)
- I/O analysis tools are invaluable in helping to navigate this complexity and to better understand I/O
  - Characterize I/O behavior of individual jobs to inform tuning decisions
  - Characterize job populations to better understand system-wide I/O stack usage and optimize deployments









## Darshan: An application I/O characterization tool for HPC





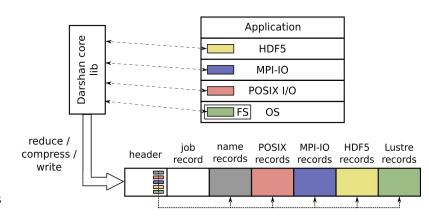
#### What is Darshan?

- Darshan is a lightweight I/O characterization tool that captures concise views of HPC application I/O behavior
  - Produces a summary of I/O activity for each instrumented job
    - Counters, histograms, timers, & statistics
    - Full I/O traces (if requested)
- Widely available
  - Deployed (and commonly enabled by default!) at many HPC facilities around the world
- Easy to use
  - No code changes required to integrate Darshan instrumentation
  - Negligible performance impact; just "leave it on"
- Modular
  - > Adding instrumentation for new I/O interfaces or storage components is straightforward



#### **How does Darshan work?**

- Darshan can insert application I/O instrumentation at link-time (for static and dynamic executables) or at runtime using LD PRELOAD (for dynamic executables)
  - > Starting in version 3.2.0, Darshan supports instrumentation of any dynamically-linked executable (MPI or not) using the LD PRELOAD method
- Darshan records file access statistics for each process as app executes
- At app shutdown, collect, aggregate, compress, and write log data
- After job completes, analyze Darshan log data
  - darshan-job-summary provides a summary PDF characterizing application I/O behavior
  - darshan-parser provides complete text-format dump of all counters in a log file
  - pydarshan Python analysis module for Darshan logs







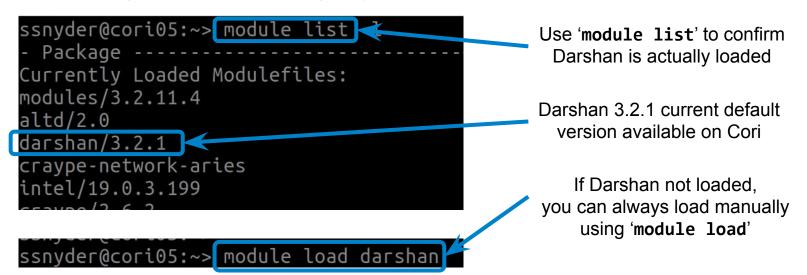






#### **MPI** applications

- Darshan is already installed and on by default on NERSC's Cori system
  - Instrumentation enabled using Cray software module that injects Darshan linker options when compiling MPI applications using Cray compiler wrappers (cc, CC, etc.)







#### **HDF5** applications

- Note that in addition to the default darshan/3.2.1 module, there is a special darshan/3.2.1-hdf5 that enables instrumentation of HDF5 APIs\*
  - Offered as a separate module to prevent non-HDF5 applications from inheriting Darshan's HDF5 library dependency

Use 'module switch' to switch to the non-default Darshan module built with HDF5 support

\*Note: this module is only compatible with cray-hdf5 versions 1.10+





- OK, Darshan is loaded...now what?
  - Just compile and run your application!
  - Darshan inserts instrumentation directly into executable
- LD PRELOAD is another option for dynamically-linked executables:
  - > This method is necessary for Python environments (i.e., mpi4py, h5py)
  - Also helpful for applications that cannot be recompiled

```
ssnyder@nid00010:/global/cscratch1/sd/ssnyder/darshan-test> module show darshan
/global/common/software/nersc/cle7/extra_modulefiles/darshan/3.2.1:

module-whatis a scalable HPC I/O characterization tool
PE_PKGCONFIG_LIBS darshan-runtime
prepend-path
prepend-path
prepend-path
prepend-path
PATH /usr/common/software/darshan/3.2.1/lib/pkoconfig
LD LIBRARY PATH /usr/common/software/darshan/3.2.1/lib
PATH /usr/common/software/darshan/3.2.1/lib

ssnyder@nid00010:/global/cscratch1/sd/ssnyder/darshan-test> export LD_PRELOAD=/usr
/common/software/darshan/3.2.1/lib/libdarshan.so
```

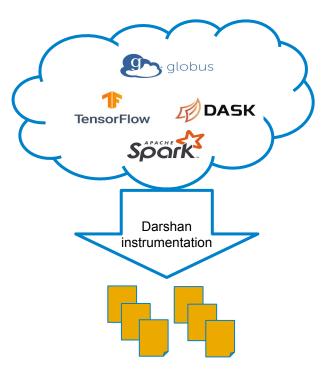
Manually set
LD\_PRELOAD to point
to Darshan's shared
library before running
your application





#### non-MPI applications

- Starting in version 3.2.0, Darshan supports instrumentation of non-MPI applications using LD\_PRELOAD (i.e., dynamically-linked binaries)
  - Users must additionally export 'DARSHAN\_ENABLE\_NONMPI=1' to enable Darshan in this case
- Note: there is a bug in the Darshan 3.2.1 module on Cori that prevents non-MPI mode from working properly
  - Users can install an explicit non-MPI version of Darshan themselves to workaround for now (e.g., 'spack install darshan-runtime~mpi+slurm')







After the application terminates, look for your log files:

Darshan logs stored in a central directory for all Cori users, use 'darshan-config --log-path' to find

```
ssnyder@cori01:~> darshan-config --log-path
/global/cscratch1/sd/darshanlogs
ssnyder@cori01:~> cd /global/cscratch1/sd/darshanlogs
ssnyder@cori01:/global/cscratch1/sd/darshanlogs>
ssnyder@cori01:/global/cscratch1/sd/darshanlogs> cd 2021/3/4
ssnyder@cori01:/global/cscratch1/sd/darshanlogs/2021/3/4> ls | grep snyder | cat ssnyder_mpi-io-test_id40245367_3-4-50083-3517743081787486417_1614894884.darshan
```

Logs further indexed using 'year/month/day' the job executed.

Log file name starts with the following pattern: 'username exename jobid...'

Note: Spack-installed Darshan versions simply put log files directly in your home directory by default -- override by setting DARSHAN\_LOG\_DIR\_PATH environment variable









After generating and locating your log, use Darshan analysis tools to inspect log file data:
Cany the log file

ssnyder@cori07:.../4> cp ssnyder\_mpi-io-test\_id40245367\_3-4-50083-35177430817874 86417\_1614894884.darshan ~/tmp-analysis/ ssnyder@cori07:.../4> cd ~/tmp-analysis/ ssnyder@cori07:~/tmp-analysis> darshan-parser ssnyder\_mpi-io-test\_id40245367\_3-4 -50083-3517743081787486417\_1614894884.darshan

Copy the log file
somewhere else for
analysis

Invoke darshan-parser (already in PATH on Cori) to get detailed counters

# <module< th=""><th><b>e&gt;</b></th><th><rank></rank></th><th><record id=""></record></th><th><counter></counter></th><th><value></value></th><th><file na<="" th=""><th>ame&gt;</th></file></th></module<>	<b>e&gt;</b>	<rank></rank>	<record id=""></record>	<counter></counter>	<value></value>	<file na<="" th=""><th>ame&gt;</th></file>	ame>
POSIX	-1	4212256	932904/85913	POSIX_OPENS	16	/global/	/CSCI
POSIX	-1	4212256	932904785913	POSIX_FILENOS	0	/global,	/csci
POSIX	-1	4212256	932904785913	POSIX_DUPS	0	/global,	/csci
POSIX	-1	4212256	932904785913	POSIX_READS	4	/global/	/csci
MPI-IO	-1	4212256	932904785913	MPIIO_INDEP_OPE	VS	8	/glo
MPI-IO	-1	4212256	932904785913	MPIIO_COLL_OPENS	S	0	/glo
MPI-IO	-1	4212256	932904785913	MPIIO_INDEP_REA	OS	4	/glo
MPI-IO	-1	4212256	932904785913	MPIIO_INDEP_WRI	TES	4	/glo

Modules use a common format for printing counters, indicating the module, rank, record ID, counter name, counter value, filename, etc. -- here sample counters are shown for both POSIX and MPI-IO modules



- But, darshan-parser output isn't so accessible for most users... use darshan-job-summary tool to produce summary PDF of app I/O behavior
  - Due to LaTeX and Perl dependencies, it may be easier just to copy Darshan logs to a personal workstation for analysis

Invoke darshan-job-summary on log file to produce PDF

```
shane@shane-x1-carbon ~/tmp-analysis $ darshan-job-summary.pl ssnyder_mpi-io-test_i d40245367_3-4-50083-3517743081787486417_1614894884.darshan shane@shane-x1-carbon ~/tmp-analysis $ shane@shane-x1-carbon ~/tmp-analysis $ ls ssnyder_mpi-io-test_id40245367_3-4-50083-3517743081787486417_1614894884.darshan ssnyder_mpi-io-test_id40245367_3-4-50083-3517743081787486417_1614894884.darshan.pdf
```

Output PDF file name based on Darshan log file name



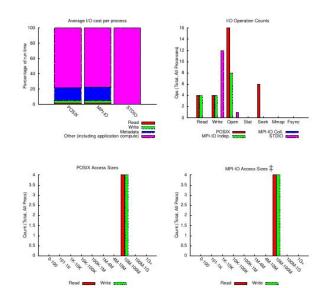


Result is a multi-page PDF containing graphs, tables, and performance estimates characterizing the I/O workload of the application

We will summarize some of the highlights in the following slides

jobid: 403177 | uid: 31074 | nprocs: 4 | runtime: 2 seconds

I/O performance estimate (at the MPI-IO layer): transferred 642 MiB at 266.40 MiB/s I/O performance estimate (at the STDIO layer): transferred 0.0 MiB at 2.08 MiB/s



Most Common Access Sizes (POSIX or MPI-IO)

access size	count
16777216	8
16777216	8
	16777216

NOTE: MPI-IO accesses are given in terms of aggregate datatype size.

File Count Summary (estimated by POSIX I/O access offsets)

(				
type	number of files	avg. size	max size	
total opened	2	33M	64M	
read-only files	0	0	0	
write-only files	1	642	642	
read/write files	1	64M	64M	
created files	2	33M	64M	





PDF header contains some high-level information on the job execution



jobid: 403177	uid: 31074	nprocs: 4	runtime: 2 seconds
Jobia. 1001//	u	IIPIOCO. ,	Tantanie. 2 decondo

I/O performance estimate (at the MPI-IO layer): transferred 642 MiB at 266.40 MiB/s
I/O performance estimate (at the STDIO layer): transferred 0.0 MiB at 2.08 MiB/s



I/O performance estimates (and total I/O volumes) provided for MPI-IO/POSIX and STDIO interfaces



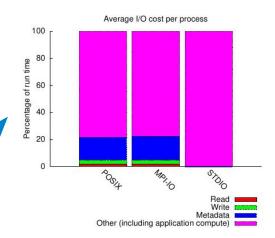


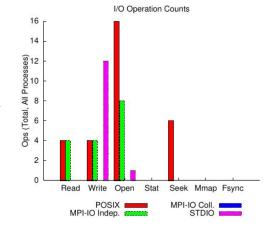
Across main I/O interfaces, how much time was spent reading, writing, doing metadata, or computing?

If mostly compute, limited opportunities for I/O tuning

What were the relative totals of different I/O operations across key interfaces?

Lots of metadata operations (open, stat, seek, etc.) could be a sign of poorly performing I/O





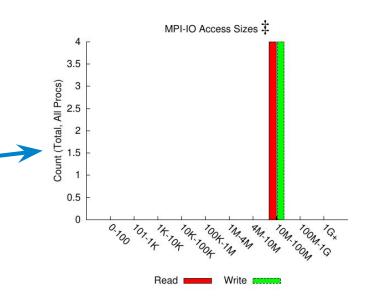




Histograms of POSIX and MPI-IO access sizes are provided to better understand general access patterns

In general, larger access sizes perform better with most storage systems

Table indicating total number of files of different types (opened, created, read-only, etc.) recorded by Darshan

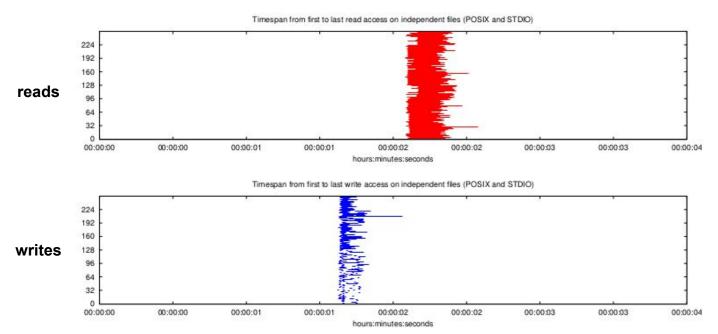


File Count Summary (estimated by POSIX I/O access offsets)

(estimated by 1 obil 1, o decess offsets)			
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write-only files	1	642	642
read/write files	1	64M	64M
created files	2	33M	64M







Darshan can also provide basic timing bounds for read/write activity, both for independent file access patterns (illustrated) or for shared file access patterns





Obtaining finer-grained details with Darshan





## Focusing analysis on individual files

- If we want to focus Darshan analysis tools on a specific file, Darshan offers a couple of different options
  - darshan-convert utility can be used to create a new Darshan log file containing a specified file record ID (obtainable from darshan-parser output)
    - e.g., 'darshan-convert --file RECORD\_ID input\_log.darshan output\_log.darshan'
    - New log file can be ran through existing log utilities we have already covered
  - darshan-summary-per-file tool can be used to generate separate job summary PDFs for every file in a given Darshan log
    - Do not use if your application opens a lot of files!

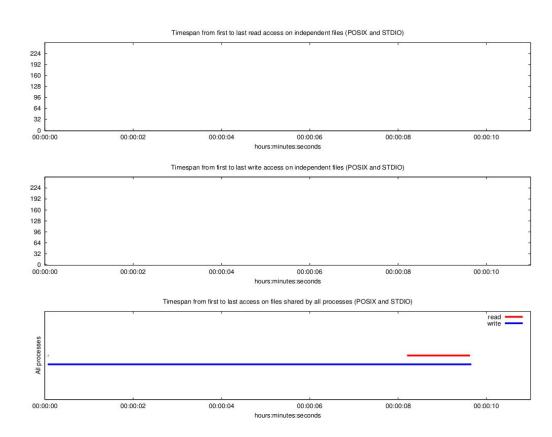




## Disabling reductions of shared records

You may notice that Darshan is unable to provide more detailed access information for shared file workloads, as illustrated here

For shared files, information from each rank is combined into a single record to save space





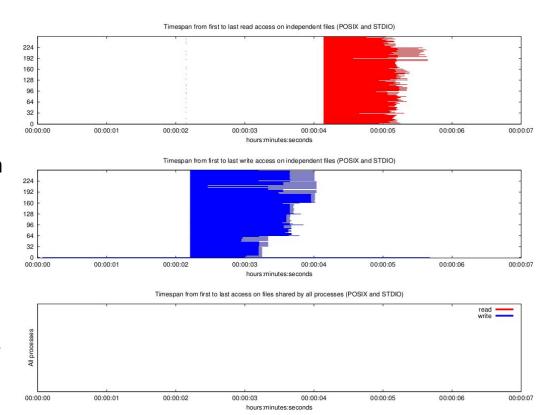


### Disabling reductions of shared records

Setting the

'DARSHAN\_DISABLE\_SHARED\_REDUCTION'
environment variable will force Darshan
to skip the shared file reduction step,
retaining each process's independent
view of access information

This results in larger log files, but may be useful in better understanding underlying access patterns in collective workloads







## Obtaining fine-grained traces with DXT

- Darshan's DXT module can be enabled at runtime for users wishing to capture detailed I/O traces for MPI-IO and POSIX interfaces
  - > Fine-grained trace data comes at cost of larger per-process memory overheads
  - > Set the DXT\_ENABLE\_IO\_TRACE environment variable to enable
- darshan-dxt-parser can be then be used to dump text-format trace data:

```
DXT POSIX module data
 DXT, file id: 11542722479531699073, file name: /qlobal/cscratch1/sd/pcarns/ior/ior.dat
 DXT, rank: 0, hostname: nid00511
 DXT, write_count: 16, read_count: 16
# DXT, mnt_pt: /global/cscratch1, fs_type: lustre
# DXT, Lustre stripe_size: 1048576, Lustre stripe_count: 24
 DXT, Lustre OST obdidx: 49 185 115 7 135 3 57 95 43 27 191 1 163 51 15 153 187 55 151 239 79
25 137 47
  Module
             Rank Wt/Rd Segment
                                              Offset
                                                                        Start(s)
                                                                                                 [OST]
                                                             Length
                                                                                       End(s)
                0 write
                                                           1048576
                                                                          0.7895
                                                                                       0.8267
                                             1048576
                                                           1048576
                                                                          0.8267
                                             2097152
                                                           1048576
                                                                          0.9843
                                                                                        1.0189
```

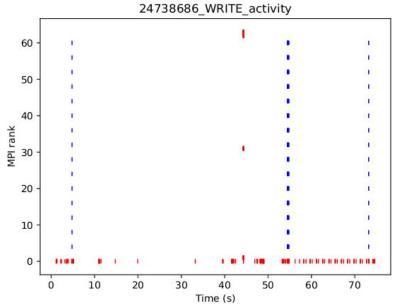




## Obtaining fine-grained traces with DXT

dxt\_analyzer Python script installed with darshan-util can be used to help visualize read/write trace activity:
24738686 WRITE activity

Provides details on each I/O operation issued by each rank, providing a complete picture of which ranks are performing I/O and how long they are spending on I/O





## **New and upcoming Darshan features: HDF5 instrumentation**





#### Available in Darshan 3.2.0+

- HDF5 offers a convenient abstraction for large data collections, but it can be difficult to understand how it interacts with lower layers of the I/O stack that most impact performance
- To help better understand HDF5 usage and performance, we have developed Darshan instrumentation modules for HDF5 file (H5F) and dataset (H5D) APIs
  - What are file and dataset properties?
  - How are datasets accessed?
  - How are datasets organized within files?
  - ➤ Do HDF5 accesses decompose efficiently to lower-level (i.e., MPI-IO and POSIX) accesses? If not, do any optimizations make sense?





#### Available in Darshan 3.2.0+

- H5F instrumentation highlights:
  - > Operation counts
    - open/create
    - flush
  - ➤ MPI-IO usage
  - Metadata timing

```
#<module>
            <rank>
                    <record id> <counter>
                                             <value> <file na
                                H5F OPENS
        11831850109748558379
                                                 /home/shane/
                                                 /home/shane/
        11831850109748558379
                                 H5F FLUSHES 0
                                 H5F USE MPIIO
        11831850109748558379
                                                     /home/sh
                                 H5F F OPEN_START_TIMESTAMP
        11831850109748558379
        11831850109748558379
                                 H5F F CLOSE START TIMESTAMP
        11831850109748558379
                                 H5F F OPEN END TIMESTAMP
        11831850109748558379
                                 H5F F CLOSE END TIMESTAMP
        11831850109748558379
                                 H5F F META TIME 0.019466
```





#### Available in Darshan 3.2.0+

- H5D instrumentation highlights:
  - Operation counts:
    - open/create
    - read/write
    - flush
  - Total bytes read/written
  - Access size histograms
  - Dataspace selection types
    - Regular hyperslab
    - Irregular hyperslab
    - Points
  - Dataspace total dimensions, points
  - Chunking parameters
  - > MPI-IO collective usage
  - Deprecated function usage
  - Read, write, and metadata timing

```
#<module>
             <rank> <record id> <counter>
                                               <value>
        7600138186531619366 H5D OPENS
                                               /home/sh
H5D -1
        7600138186531619366 H5D READS
                                               /home/sl
        7600138186531619366 H5D WRITES
H5D - 1
                                               /home/sh
H5D - 1
        7600138186531619366 H5D FLUSHES 0
                                               /home/s
        7600138186531619366 H5D BYTES READ
H5D -1
                                               4194304
        7600138186531619366 H5D BYTES WRITTEN
H5D-1
                                                   419
H5D - 1
        7600138186531619366 H5D RW SWITCHES 4
                                                   /hoi
        7600138186531619366 H5D REGULAR HYPERSLAB SI
H<sub>5</sub>D
xt4
H<sub>5</sub>D
        7600138186531619366 H5D_IRREGULAR_HYPERSLAB_
xt4
H5D -1
        7600138186531619366 H5D_POINT_SELECTS
        7600138186531619366 H5D MAX READ TIME SIZE
H5D -1
        7600138186531619366 H5D_MAX_WRITE_TIME_SIZE
H5D -1
         7600138186531619366 H5D SIZE READ AGG 0
```





#### Available in Darshan 3.2.0+

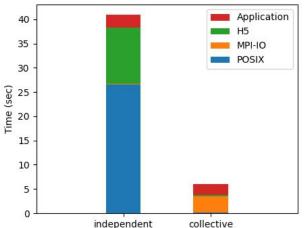
- Using the MACSio¹ HDF5 plugin, run a couple of simple examples demonstrating the types of insights HDF5 I/O instrumentation can enable
  - > 60-process (5-node) single shared file, 3d mesh, write roughly 1 GiB of cumulative H5D data
  - Compare performance of collective and independent I/O configurations

b/w: ~30 MB/sec

POSIX I/O dominates, H5 incurs non-negligible overhead forming this workload

Negligible time spent in

Average per-process time spent in I/O



b/w: ~290 MB/sec

H5 and POSIX incur minimal overhead for this workload

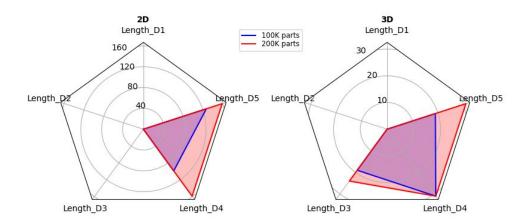
MPI-IO collective I/O algorithm dominates





#### Available in Darshan 3.2.0+

- Using the MACSio¹ HDF5 plugin, run a couple of simple examples demonstrating the types of insights HDF5 I/O instrumentation can enable
  - > 60-process (5-node) single shared file, 3d mesh, write roughly 1 GiB of cumulative H5D data
  - Compare performance of collective and independent I/O configurations



Radar plots, or other methods, can be used to help visualize characteristics of HDF5 dataset accesses

Dataset access patterns could be used to help set/optimize chunking parameters to limit accesses to as few chunks as possible





## New and upcoming Darshan features: DAOS instrumentation





#### **DAOS** instrumentation

#### Work in progress

- Intel's DAOS offers an exciting new storage paradigm for HPC apps, utilizing object-based storage interfaces over a combo of SCM and SSD devices
  - ➤ libdfs: DAOS's POSIX file system emulation API
  - libdaos: DAOS's native object (key-val) API
- Darshan will instrument libdaos and libdfs APIs to help provide insight into application and I/O middleware usage of DAOS
  - Legacy POSIX app usage and performance characteristics
  - Usage and performance characteristics of libdaos users (libdfs, HDF5 VOL, MPI-IO, etc.)

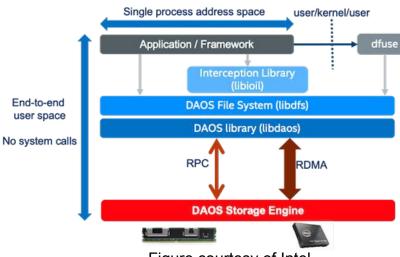


Figure courtesy of Intel



#### **DAOS** instrumentation

#### Work in progress

- libdfs instrumentation highlights:
  - Operation counts:
    - open
    - read/write
    - punch
    - stat
  - ➤ Total bytes read/written
  - Access size histograms
  - > File chunk size
  - DTX usage (strict consistency mode)
  - Corresponding DAOS object record ID
    - Necessary to link Darshan's DFS records with native DAOS records
  - Read, write, and metadata timing

```
#<module>
            <rank> <record id> <counter>
                                             <value>
DFS 0
                                             c3bcfc9a
        8492698188889325809 DFS OPENS
DFS 0
        8492698188889325809 DFS GLOBAL OPENS
DFS 0
        8492698188889325809 DFS LOOKUPS 1
                                             c3bcfc9a
DFS 0
        8492698188889325809 DFS DUPS
                                             c3bcfc9a
DFS 0
        8492698188889325809 DFS READS
                                             c3bcfc9a
        8492698188889325809 DFS READXS
DFS 0
                                             c3bcfc9a
DFS 0
        8492698188889325809 DFS WRITES 0
                                             c3bcfc9a
DFS 0
        8492698188889325809 DFS WRITEXS 0
                                             c3bcfc9a
        8492698188889325809 DFS NB READS
DFS 0
                                                 c3bc
DFS 0
        8492698188889325809 DFS NB WRITES
                                                 c3bc
DFS 0
        8492698188889325809 DFS GET SIZES
                                                 c3bc
DFS 0
        8492698188889325809 DFS_PUNCHES 0
                                             c3bcfc9a
```

**Note**: Unsurprisingly, Darshan's DFS instrumentation module closely follows the design of the POSIX module

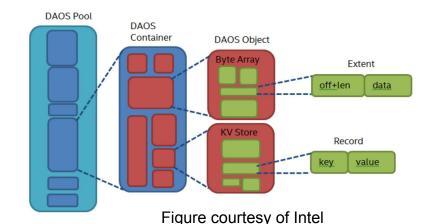




#### **DAOS** instrumentation

#### Work in progress

- libdaos instrumentation highlights:
  - Operation counts:
    - open
    - fetch (read) and update (write)
    - list (enumeration)
    - punch
  - Total bytes read/written
  - Access size histograms
  - Object class parameters
    - layout (static or dynamic striping)
    - redundancy (replication, erasure coding)
  - Container/pool UUIDs
  - > Read, write, and metadata timing



Note: Each DAOS object is multi-level key-val store, creating challenges for deciding what can be instrumented using a fixed set of Darshan counters -- we are still investigating what characteristics about key access patterns to capture





New and upcoming Darshan features: pydarshan





## pydarshan log file analysis

#### Available in Darshan 3.3.0, coming soon!

- \* Thanks to **Jakob Luettgau (DKRZ)** for contributing the majority of the pydarshan code, examples, and documentation
- Darshan has traditionally offered only the C-based darshan-util library and a handful of corresponding utilities to users
  - > Development of custom Darshan analysis utilities is cumbersome, requiring users to either:
    - Develop analysis tools in C using the low-level darshan-util library
    - Perform an inconvenient conversion from darshan-parser text output
- pydarshan has been developed\* to simplify the interfacing of analysis tools with Darshan log data
  - > Use Python CFFI module to provide Python bindings to the native darshan-utils C API
  - > Expose Darshan log data as dictionaries, pandas dataframes, and numpy arrays
- We are hopeful pydarshan will lead to a richer ecosystem for Darshan log analysis utilities





## pydarshan log file analysis

#### Available in Darshan 3.3.0, coming soon!

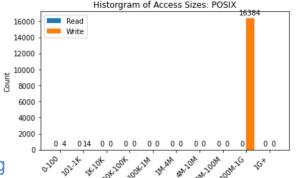
We've already found Jupyter notebooks to be an effective way of sharing pydarshan analysis examples (code, documentation, visualizations) with users, collaborators, etc.

```
In [1]: import darshan
        report = darshan.DarshanReport("example-logs/example.darshan", read all=True) # Default
        report.info()
                        example-logs/example.darshan
        Filename:
                        2017-03-20 04:07:47 to 2017-03-20 04:09:43 (Duration 0:01:56)
        Times:
                        /global/project/projectdirs/m888/glock/tokio-abc-results/bin.edison/vpici
        Executeable:
        rs/glock/tokioabc-s.4478544/vpicio/vpicio.hdf5 32
        Processes:
                        2048
        JobID:
                        4478544
        UID:
                        69615
        Modules in Log: ['POSIX', 'MPI-IO', 'LUSTRE', 'STDIO']
        Loaded Records: {'POSIX': 1, 'MPI-IO': 1, 'STDIO': 1, 'LUSTRE': 1}
        Name Records:
        Darshan/Hints: {'lib ver': '3.1.3', 'h': 'romio no indep rw=true;cb nodes=4'}
        DarshanReport: id(140124449925824) (tmp)
```

```
In [3]: # access histograms
plt = plot_access_histogram(report, 'POSIX')
plt.show()

Summarizing... iohist POSIX
```

Summarizing... iohist POSIX



In just a few lines of code, users can read a Darshan log into memory and generate plots describing access patterns





## pydarshan log file analysis

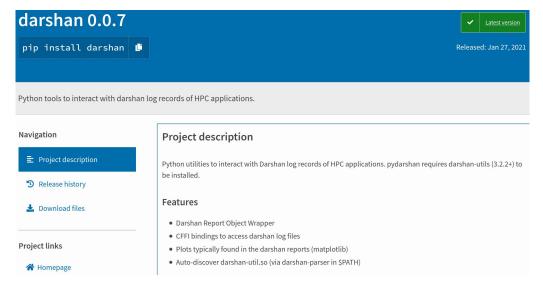
#### Available in Darshan 3.3.0, coming soon!

A beta version of pydarshan is currently available in PyPI and ready for users

to experiment with

Use 'pip install darshan' to install the pydarshan module from PyPI on your system

Alternatively, pydarshan can be installed alongside darshan-util, by providing '--enable-pydarshan --enable-shared' configure options







## Other upcoming Darshan features

- In addition to those covered today, the following features are also on the Darshan roadmap:
  - > Autoperf instrumentation module (available in Darshan 3.3.0)
    - APMPI MPI communication counters
    - CrayXC compute and network counters for Cray XC systems
    - Thanks to Sudheer Chunduri (ALCF) for this contribution!
  - > PnetCDF instrumentation module (work in progress)
    - Full instrumentation of PnetCDF blocking/nonblocking APIs
    - Thanks to Claire Lee (NWU) and Wei-keng Liao (NWU) for this contribution!
  - Fork handlers for non-MPI mode (work in progress, available in 'dev-fork-safe' branch)
    - Numerous updates to allow forked processes to generate their own Darshan logs.
  - ➤ Enhanced analysis tools and report generation (???)
    - Building off of our pydarshan log utility bindings, we want to revamp our analysis tools and report generation using Python





## Broader system-wide analysis using Darshan: Gauge



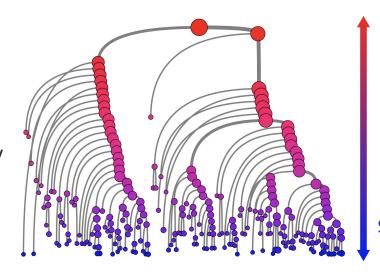


Special acknowledgement for the Gauge tool and the content of this slide goes to **Mihailo Isakov**, et al. (**TAMU**)

## Gauge

#### I/O clustering and analysis tool for Darshan logs

- Allow system administrators and I/O experts to gain insights into classes of I/O workloads running on a computer
- HDBSCAN clustering used to organize Darshan logs into hierarchy of clusters that exhibit similar I/O characteristics
- Gauge users may select clusters to see more information on the clusters I/O characteristics
  - User and application details
  - Job features
  - Access patterns



Larger, sparser clusters

Node height shows cluster density

Smaller, denser clusters

Iskaov et al. "HPC I/O Throughput Bottleneck Analysis with Explainable Local Models", SC'20

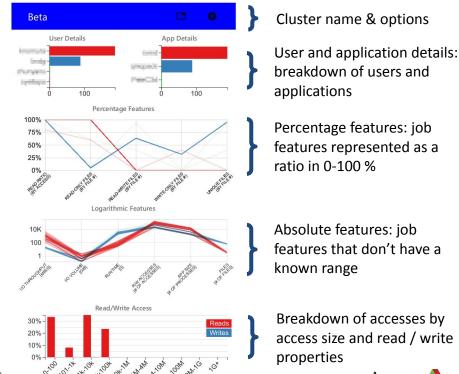
del Rosario et al. "Gauge: An Interactive Data-Driven Visualization Tool for HPC Application I/O Performance Analysis", PDSW'20

## Special acknowledgement for the Gauge tool and the content of this slide goes to **Mihailo Isakov**, et al. (**TAMU**)

## Gauge

#### I/O clustering and analysis tool for Darshan logs

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## Wrapping up

- We've covered a range of topics today demonstrating the role Darshan can play in developing a better understanding of HPC I/O behavior
  - > Runtime characterization of application usage of a number of different I/O libraries
    - Traditional I/O libraries like POSIX, MPI-IO, HDF5, as well as upcoming libraries such as DAOS
  - I/O insights enabled via Darshan data, using existing utilities in the Darshan repo or using 3rd party tools like Gauge
    - pydarshan can greatly simplify the process of extending existing Darshan analysis tools or developing custom analysis tools
- Please reach out with any questions, comments, or feedback!
- Darshan website: <a href="https://www.mcs.anl.gov/research/projects/darshan/">https://www.mcs.anl.gov/research/projects/darshan/</a>
- Darshan-users mailing list: <u>darshan-users@lists.mcs.anl.gov</u>
- Source code, issue tracking: <a href="https://xgitlab.cels.anl.gov/darshan/darshan">https://xgitlab.cels.anl.gov/darshan/darshan</a>
- NERSC documentation: <a href="https://docs.nersc.gov/development/performance-debugging-tools/darshan">https://docs.nersc.gov/development/performance-debugging-tools/darshan</a>











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